

On limit theorems for continued fractions

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Abstract

It is shown that for sums of functionals of digits in continued fraction expansion the Kolmogorov-Feller weak laws of large numbers and the Khinchine-Lévy-Feller-Raikov characterization of the domain of attraction of the normal law hold.

Key words: weak law of large numbers, domain of attraction of the normal law, ψ -mixing, continued fractions.

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1 Introduction and result

Let $a_n(x)$, $n \in \mathbb{N} = \{1, 2, \dots\}$, denote the partial quotients (or digits) in the simple non-terminating continued fraction expansion of an irrational number $x \in (0, 1]$, (Cf. [20])

$$x = \frac{1}{a_1(x) + \frac{1}{a_2(x) + \frac{1}{a_3(x) + \dots}}} = \frac{1}{a_1(x) +} \frac{1}{a_2(x) +} \frac{1}{a_3(x) +} \dots$$

Let \mathcal{B} denote Borel subsets of $(0, 1]$ and P denote the Gauss' measure

$$P(A) = \frac{1}{\ln 2} \int_A \frac{dx}{1+x}, \quad A \in \mathcal{B}.$$

It is well-known that the random sequence $\{a_n\}$ defined on the probability space $((0, 1], \mathcal{B}, P)$ is strictly stationary.

The literature concerning the limit theory for functionals of digits of continued fraction expansion (see e.g. [20], [14], [16], [17], [26], [2]) reveals that some classical results for sums of i.i.d. random variables (see e.g. [9], [25], [14], [30]) found their full analogies in this theory. One example is the Marcinkiewicz-Zygmund law of large numbers (Cf. [23], [27]), or more generally the complete convergence (Cf. [33], [26]).

In this note we discover further analogies. The first one is motivated by Lemma in [7] (see also Theorem 4.13 in [30]) and §4, Ch. VI in [23] (see also Satz XII in [22] & [21], Theorem 1 in §7, Ch. VII in [8], Theorem 1.3 in [10]).

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Theorem 1 *Let $c_n \rightarrow \infty$ be a sequence of positive numbers and f be a Borel function. In order that there exist a sequence $\{b_n\}$ such that*

$$c_n^{-1} \left(\sum_{k=1}^n f(a_k) - b_n \right) \rightarrow_{\mathbb{P}} 0 \quad (1.1)$$

it is necessary and sufficient that simultaneously

$$n\mathbb{P}[|f(a_1)| > c_n] \rightarrow 0, \quad \frac{n}{c_n^2} \mathbb{E}[f^2(a_1) I_{|f(a_1)| \leq c_n}] \rightarrow 0. \quad (1.2)$$

If the latter conditions are satisfied, we can set $b_n = n\mathbb{E}[f(a_1) I_{|f(a_1)| \leq c_n}]$.

Note that by the well-known formula

$$x^2 \mathbb{P}[|Z| > x] + \mathbb{E}[|Z|^2 I_{|Z| \leq x}] = 2 \int_0^x y \mathbb{P}[|Z| > y] dy, \quad (1.3)$$

(1.2) is equivalent to

$$\int_0^1 n \mathbb{P}[|f(a_1)| > \sqrt{x} c_n] dx \rightarrow 0. \quad (1.4)$$

If there exists some additional knowledge on c_n then (1.2) can be weakened.

Theorem 2 *Suppose that c_n is a sequence of positive numbers such that*

$$\underline{\lim}_n \frac{n}{c_n^2} (c_{n+1}^2 - c_n^2) > 1. \quad (1.5)$$

Then for any Borel function f

$$(c_n^{-1} \sum_{k=1}^n (f(a_k) - \mathbb{E}[f(a_k) I_{|f(a_k)| \leq c_n}]) \rightarrow_{\mathbb{P}} 0) \iff (n\mathbb{P}[|f(a_1)| > c_n] \rightarrow_n 0). \quad (1.6)$$

Theorem 2 and Theorem 1.9.8 in [3] yield

Corollary 1 *Suppose that f is a Borel function, $c_n = n^{\frac{1}{r}} h(n)$, where h is a slowly varying function in the sense of Karamata and $r \in (0, 2)$. Then the relation (1.6) holds.*

Let c_n denote the accumulated entrance fees up to the n -th trial in the St. Petersburg game, i.e. $c_n = n \log_2 n$ (Cf. [7]). Since $n \ln 2 \mathbb{P}[a_1 > n] \sim 1$, therefore $\mathbb{E}[a_1 I_{a_1 \leq c_n}] \sim \log_2 n$ and by Corollary 1 (see also Theorem 3 in [35]) we get for $f(x) = x$

$$c_n^{-1} \sum_{k=1}^n a_k \rightarrow_{\mathbb{P}} 1.$$

On the other hand, by Theorem 4 in [35], the convergence in probability cannot be replaced by the almost sure one.

The second analogy is motivated by the famous characterization of the domain of attraction of the normal law due to Khinchine (Cf. [19]), Lévy (Cf. [24]) and Feller (Cf. [6]) (see also Corollary 1, §5, Ch. XVII in [8])

Theorem 3 *Let f be a Borel function. In order that there exist sequences $\{c_n\}$ and $\{d_n\}$ such that $\lim_{n \rightarrow \infty} \mathcal{L}(c_n^{-1}(\sum_{k=1}^n f(a_k) - d_n)) = \mathcal{N}(0, 1)$ it is necessary and sufficient that the function $E[f^2(a_1)I_{[|f(a_1)| \leq x]}]$ is slowly varying in the sense of Karamata. If the latter condition is satisfied, we can set $d_n = nE[f(a_1)]$.*

Theorem 3 improves Proposition 2.9 (II) in [32] obtained under additional assumptions on c_n and d_n . Furthermore, we no longer have to assume that condition (*) on page 56 of [32] is satisfied. By Theorem 3 and the results in [35] we get that functionals of digits in continued fraction expansion satisfy the Raikov principle (Cf. Twierdzenie 4, §28, Ch. V in [9]), namely

Corollary 2 *Suppose that f is a Borel function and c_n is a positive sequence. Then $c_n^{-2} \sum_{k=1}^n f^2(a_k) \rightarrow_{\mathbb{P}} 1$ and $E[f(a_1)] = 0$ if and only if*

$$\lim_{n \rightarrow \infty} \mathcal{L}(c_n^{-1} \sum_{k=1}^n f(a_k)) = \mathcal{N}(0, 1).$$

As it can be concluded from the rest of this note, the presented results remain true in a little bit more general mathematical environment.

2 Preliminaries

We group here different results that will be used later on. Let $\{X_k\}_{k \in \mathbb{N}}$ be a random sequence defined on a probability space (Ω, \mathcal{F}, P) , denote by $\{\tilde{X}_k\}$ it's independent copy and in the case of stationarity by $\{X_k^*\}$ it's i.i.d. associated sequence (all sequences are sharing the same probability space). Define

$$S_n = \sum_{k=1}^n X_k, \quad \hat{X}_k = X_k - \tilde{X}_k, \quad \hat{S}_n = \sum_{k=1}^n \hat{X}_k,$$

$$M_n = \max_{1 \leq k \leq n} |X_k|, \quad \widehat{M}_n = \max_{1 \leq k \leq n} |\hat{X}_k|, \quad M_n^* = \max_{1 \leq k \leq n} |X_k^*|.$$

Denote by \mathcal{F}_k^m the σ -field generated X_k, X_{k+1}, \dots, X_m , $m \in \mathbb{N}$, and recall the following coefficients of dependence

$$\psi_n = \sup_{k \in \mathbb{N}} \sup \left\{ \left| \frac{P(A \cap B)}{P(A)P(B)} - 1 \right|; P(A)P(B) > 0, A \in \mathcal{F}_1^k, B \in \mathcal{F}_{n+k}^\infty \right\};$$

$$\psi_n^* = \sup_{k \in \mathbb{N}} \sup \left\{ \frac{P(A \cap B)}{P(A)P(B)}; P(A)P(B) > 0, A \in \mathcal{F}_1^k, B \in \mathcal{F}_{n+k}^\infty \right\};$$

$$\psi'_n = \inf_{k \in \mathbb{N}} \inf \left\{ \frac{P(A \cap B)}{P(A)P(B)}; P(A)P(B) > 0, A \in \mathcal{F}_1^k, B \in \mathcal{F}_{n+k}^\infty \right\};$$

$$\varphi_n = \sup_{k \in \mathbb{N}} \sup \{|P(B|A) - P(B)|; P(A) > 0, A \in \mathcal{F}_1^k, B \in \mathcal{F}_{n+k}^\infty\};$$

$$\rho_n = \sup_{k \in \mathbb{N}} \sup \{|Corr(f, g)|; f \in L_{\text{real}}^2(\mathcal{F}_1^k), g \in L_{\text{real}}^2(\mathcal{F}_{n+k}^\infty)\}.$$

It is well-known that

$$\psi_n = \max\{\psi_n^* - 1, 1 - \psi_n'\}, \psi_n \geq 2\varphi_n, \psi_n \geq \rho_n, 1 - \varphi_n \geq \psi_n', 4\varphi_n \geq (\rho_n)^2, \quad (2.7)$$

for every $n \geq 1$ (Cf. [5], p.109). By Theorem 5.2 in [5] the symmetrized $\psi_n, \psi_n^*, \psi_n', \varphi_n, \rho_n$ coefficients of the sequence $\{\widehat{X}_k\}$, say $\widehat{\psi}_n, \widehat{\psi}_n^*, \widehat{\psi}_n', \widehat{\varphi}_n, \widehat{\rho}_n$, satisfy

$$\widehat{\psi}_n \leq (1 + \psi_n)^2 - 1, \widehat{\psi}_n^* \leq (\psi_n^*)^2, \widehat{\psi}_n' \geq (\psi_n')^2, \widehat{\varphi}_n \leq 1 - (1 - \varphi_n)^2, \widehat{\rho}_n \leq \rho_n, \quad (2.8)$$

for every $n \geq 1$.

The following lemma is a dependent version of Lévy's inequality.

Lemma 1 *Assume that $\mathcal{L}(S_n - S_k)$ are symmetric for $n > k \geq 1$. Then for $n \geq 1$ and $x > 0$*

$$2P[|S_n| > x] \geq \psi_1' P[\max_{1 \leq k \leq n} |S_k| > x].$$

PROOF OF LEMMA 1

Consider the sets

$$C_1^+ = [S_1 > x], \quad C_k^+ = [|S_1| \leq x, |S_2| \leq x, \dots, |S_{k-1}| \leq x, S_k > x],$$

$$C_1^- = [-S_1 > x], \quad C_k^- = [|S_1| \leq x, |S_2| \leq x, \dots, |S_{k-1}| \leq x, -S_k > x],$$

and set $C_k = C_k^+ \cup C_k^-$, $C = \bigcup_{k=1}^n C_k$. Since the C_k are disjoint we obtain

$$\begin{aligned} 2P[|S_n| > x] &= 2P[\max_{1 \leq k \leq n} |S_k| > x, |S_n| > x] \\ &= 2 \sum_{k=1}^n P(C_k \cap [|S_n| > x]) \\ &\geq 2 \sum_{k=1}^n (P(C_k^+ \cap [S_n > x]) + P(C_k^- \cap [-S_n > x])) \\ &\geq 2 \sum_{k=1}^n (P(C_k^+ \cap [S_n - S_k \geq 0]) + P(C_k^- \cap [-S_n + S_k \geq 0])) \\ &\geq \psi_1' \sum_{k=1}^n (P(C_k^+) + P(C_k^-)) = \psi_1' \sum_{k=1}^n P(C_k) \\ &= \psi_1' P(C) = \psi_1' P[\max_{1 \leq k \leq n} |S_k| > x], \end{aligned}$$

because $2P[S_n - S_k \geq 0] = 2P[S_n - S_k \leq 0] = 1 + P[S_n - S_k = 0] \geq 1$. This proves Lemma 1. \square

It is well-known that if $E|X| < \infty$, $E|Y| < \infty$, and X is \mathcal{F}_1^k measurable while Y is \mathcal{F}_{n+k}^∞ measurable then

$$|E[XY] - E[X]E[Y]| \leq \psi_n E|X|E|Y| \quad (2.9)$$

(Cf. Lemma 1.2.11 in [26]).

The statement below is a dependent version of Kolmogorov's inverse inequality (Cf. [25], p.235, [1], Theorem 2.8).

Lemma 2 Let $\{X_k\}$ be a strictly stationary sequence such that $|X_k| \leq c$ and $\mathcal{L}(S_k)$ are symmetric. Then for every $x > 0$

$$P\left[\max_{1 \leq k \leq n} |S_k| > x\right] \geq \frac{\psi'_1(E[S_n^2] - x^2)}{4(1 + \psi_1)E[S_n^2] + \psi'_1(2(c+x)^2 - x^2)}, \quad n \geq 1.$$

PROOF OF LEMMA 2

Let C_k and C be as in the proof of Lemma 1. We have

$$E[S_n^2 I_{C_k}] \leq 2E[(S_n - S_k)^2 I_{C_k}] + 2E[S_k^2 I_{C_k}].$$

By (2.9) we get

$$E[(S_n - S_k)^2 I_{C_k}] \leq (1 + \psi_1)E[(S_n - S_k)^2]P(C_k).$$

Since $|S_k I_{C_k}| \leq (c+x)I_{C_k}$ so by stationarity and Lemma 1 we obtain ($S_0 = 0$)

$$\begin{aligned} E[S_n^2 I_C] &\leq 2 \sum_{k=1}^n ((1 + \psi_1)E[(S_n - S_k)^2]P(C_k) + (c+x)^2 P(C_k)) \\ &\leq 2 \sum_{k=1}^n P(C_k) ((1 + \psi_1) \max_{1 \leq k \leq n} E[(S_n - S_k)^2] + (c+x)^2) \\ &= 2 \sum_{k=1}^n P(C_k) ((1 + \psi_1) \max_{1 \leq k \leq n} E[S_k^2] + (c+x)^2) \\ &\leq 2 \sum_{k=1}^n P(C_k) ((1 + \psi_1)E[\max_{1 \leq k \leq n} S_k^2] + (c+x)^2) \\ &\leq 2 \sum_{k=1}^n P(C_k) \left(\frac{2(1 + \psi_1)}{\psi'_1} E[S_n^2] + (c+x)^2 \right) \\ &= 2P(C) \left(\frac{2(1 + \psi_1)}{\psi'_1} E[S_n^2] + (c+x)^2 \right). \end{aligned}$$

On the other hand

$$E[S_n^2 I_C] = E[S_n^2] - E[S_n^2 I_{\Omega \setminus C}] \geq E[S_n^2] + x^2 P(C) - x^2.$$

This completes the proof. \square

The next inequality follows from the proof on p.298 in [29].

Proposition 1 Suppose $\{X_k\}$ is a strictly stationary sequence and $\varphi_m < 1$. Then for every $x \geq 0$ and every $n \geq m \geq 1$

$$(1 - \varphi_m)P[M_{[n/m]}^* > x] \leq P[M_n > x] \leq m(1 + \varphi_m)P[M_{[n/m]+1}^* > x].$$

We will need the following estimate

Lemma 3 Suppose $\{X_k\}$ is a strictly stationary sequence and $\psi'_1 > 0$. Then for every $x \geq 0$ and every $n \geq m \geq 1$

$$(1 - \widehat{\varphi}_1)\widehat{\psi}'_1(1 - e^{-nP[\widehat{X}_1 > 2x]}) \leq 4mP[|\widehat{S}_{[n/m]+1}| > x].$$

PROOF OF LEMMA 3

It is easy to see that for every $x \geq 0$ we have

$$P[M_n > 2x] \leq P[\max_{1 \leq k \leq n} |S_k| > x]. \quad (2.10)$$

Therefore by Proposition 1 and Lemma 1 we obtain

$$\begin{aligned} (1 - \widehat{\varphi}_1) \widehat{\psi}'_1 (1 - e^{-nP[\widehat{X}_1 > 2x]}) &\leq (1 - \widehat{\varphi}_1) \widehat{\psi}'_1 P[\max_{1 \leq k \leq n} |\widehat{X}_k^*| > 2x] \\ &\leq \widehat{\psi}'_1 P[\widehat{M}_n > 2x] \leq 2m \widehat{\psi}'_1 P[\widehat{M}_{\lfloor n/m \rfloor + 1} > 2x] \\ &\leq 2m \widehat{\psi}'_1 P[\max_{1 \leq k \leq \lfloor n/m \rfloor + 1} |\widehat{S}_k| > x] \leq 4m P[|\widehat{S}_{\lfloor n/m \rfloor + 1}| > x]. \end{aligned}$$

This is our assertion. \square

The following estimates are consequences of Lemma 2.1 and 2.3 in [4].

Proposition 2 *Suppose $\rho_1 < 1$ and $\sum_{n=1}^{\infty} \rho_{2^n} < \infty$ for a strictly stationary sequence $\{X_k\}$ with finite variances. Then there exist positive constants C, D depending only on $\{\rho_k\}$ such that $Cn\text{Var}[X_1] \leq \text{Var}[S_n] \leq Dn\text{Var}[X_1]$, $n \geq 1$.*

The proof of the next statement is an easy consequence of the hint on p.91 in [1] and it is included here for the reader convenience.

Lemma 4 *Suppose that $\{X_k\}$ is a strictly stationary sequence such that $\psi'_1 > 0$ and $\mathcal{L}(c_n^{-1}(S_n - b_n))$ are asymptotically normal for $c_n^2 = nh(n)$ where $h(n)$ is a slowly varying sequence. Then $\sup_n c_n^{-q} E|\widehat{S}_n|^q < \infty$, for any $q \in (0, 2)$.*

PROOF OF LEMMA 4

Suppose $d > 0$ and $\delta \in (0, \frac{1}{2} \widehat{\psi}'_1 (1 - \widehat{\varphi}_1))$ are such that

$$P[c_n^{-1} |\widehat{S}_n| > d] \leq \delta,$$

for $n > N_\delta$. By Lemma 1 we have for $n > N_\delta$

$$P[c_{mn}^{-1} \max_{1 \leq k \leq m} |\widehat{S}_{nk} - \widehat{S}_{n(k-1)}| > d] \leq \frac{2}{\widehat{\psi}'_1} P[c_{mn}^{-1} |\widehat{S}_{mn}| > d] \leq \frac{2\delta}{\widehat{\psi}'_1}.$$

On the other hand by Proposition 1

$$\begin{aligned} P[c_{mn}^{-1} \max_{1 \leq k \leq m} |\widehat{S}_{nk} - \widehat{S}_{n(k-1)}| > d] \\ &\geq (1 - \widehat{\varphi}_1) P[c_{mn}^{-1} \max_{1 \leq k \leq m} |\widehat{S}_{n,k}^*| > d] \\ &= (1 - \widehat{\varphi}_1) (1 - (1 - P[c_{mn}^{-1} |\widehat{S}_n| > d])^m), \end{aligned}$$

where $\widehat{S}_{n,k}^*$ are independent copies of $\widehat{S}_{nk} - \widehat{S}_{n(k-1)}$. Thus

$$P[c_{mn}^{-1} |\widehat{S}_n| > d] \leq 1 - (1 - \frac{2\delta}{\widehat{\psi}'_1 (1 - \widehat{\varphi}_1)})^{\frac{1}{m}}.$$

From

$$1 - (1 - a)^{\frac{1}{m}} \sim \frac{1}{m} \ln \frac{1}{1 - a}, \quad m \rightarrow \infty, \quad 0 < a < 1,$$

it follows that there exists a constant $A = A(\delta, \widehat{\psi}'_1, \widehat{\varphi}_1, N_\delta)$ such that

$$P[c_{mn}^{-1}|\widehat{S}_n| > d] \leq \frac{A}{m}. \quad (2.11)$$

Since $c_n^2 = nh(n)$, where $h(n)$ is a slowly varying sequence so by the Uniform Convergence Theorem for R (Cf. [3], Theorem 1.5.2, p.22) we have that for every $\gamma > 0$ there exists $N_\gamma > N_\delta$ such that

$$\left| \frac{m^{-\gamma} n^{-\gamma} h(mn)}{n^{-\gamma} h(n)} - m^{-\gamma} \right| = \left| \frac{c_{mn}}{m^{\frac{1}{2}+\gamma} c_n} - \frac{1}{m^\gamma} \right| < \frac{1}{2},$$

for $n \geq N_\gamma$ uniformly in $m \geq 1$. Thus

$$\frac{c_{mn}}{c_n} < m^{\frac{1}{2}+\gamma} (m^{-\gamma} + \frac{1}{2}) \leq \frac{3}{2} m^{\frac{1}{2}+\gamma},$$

for $n \geq N_\gamma$, $m \geq 1$ which with (2.11) gives

$$mP[c_n^{-1}|\widehat{S}_n| > \frac{3}{2} dm^{\frac{1}{2}+\gamma}] \leq A,$$

for $n \geq N_\gamma$, $m \geq 1$. Replacing m with $m^{\frac{2}{1+2\gamma}}$ we obtain

$$m^{\frac{2}{1+2\gamma}} P[c_n^{-1}|\widehat{S}_n| > md'] \leq A',$$

where $A' = A'(A, \gamma)$, $d' = d'(d, \gamma)$ and $n \geq N_\gamma$, $m \geq 1$. Now, taking γ such that $0 < \gamma < \frac{1}{q} - \frac{1}{2}$ we get

$$m^{-1+q} P[c_n^{-1}|\widehat{S}_n| > md'] \leq \frac{A'}{m^{1+c}},$$

where $c = \frac{2}{1+2\gamma} - q > 0$. Thus

$$\sum_{m=1}^{\infty} m^{q-1} P[c_n^{-1}|\widehat{S}_n| > md'] < \infty,$$

for $n \geq N_\gamma$. This completes the proof. \square

The next inequality follows from Proposition 1 and Proposition 4.3, Lemma 4.2 in [12].

Proposition 3 *Suppose that $\{X_k\}$ is a strictly stationary random sequence such that $\phi_1 < 1$ and $\sup\{x; P[|X_1| \leq x] < 1\} = \infty$. Then for every $q, x > 0$*

$$(1 - \phi_1) \frac{nE[|X_1|^q I_{[|X_1| > x]}]}{1 + nP[|X_1| > x]} \leq E[M_n^q], \quad n \geq 1.$$

The lemma below is a dependent analog of Khinchine's inequality (Cf. [1], p.176).

Lemma 5 *Suppose that $\{X_k\}$ is a random sequence such that $|X_k| < c$ and $0 < \psi'_1 \leq \psi_1^* < \infty$. Then for any $q \in [1, p)$ there exists a constant B_{pq} depending only on ψ'_1, ψ_1^*, p, q such that*

$$E|\widehat{S}_n|^p \leq B_{pq} \max\{E^{\frac{p}{q}}|\widehat{S}_n|^q, c^p\}, \quad n \geq 1.$$

PROOF OF LEMMA 5

From the Nagaev generalization of the inequality (3.3) in [13] (Cf. the relation (11) and the proof of Lemma in [28]) it follows that for any random sequence $\{X_k\}$ such that $|X_k| < c$ we have for $t \geq c$

$$P[\max_{1 \leq k \leq n} |S_k| > 4t] \leq \psi_1^* (P[\max_{1 \leq k \leq n} |S_k| > t])^2. \quad (2.12)$$

In view of this and Lemma 1 we get for $t > 2c$

$$P[|\widehat{S}_n| > 4t] \leq (\gamma P[|\widehat{S}_n| > t])^2,$$

where $\gamma = \frac{2\sqrt{\widehat{\psi}_1^*}}{\psi'_1}$. If we set

$$t_0 = 4^{\frac{1}{q}} \gamma^2 \max\{E^{\frac{1}{q}}|\widehat{S}_n|^q, c\} > 2c$$

then by the Markov inequality we obtain for $t \geq t_0$

$$\gamma^2 P[|\widehat{S}_n| > t] \leq \frac{1}{4}.$$

From this we have

$$\begin{aligned} E|\widehat{S}_n|^p &= p \int_0^{4t_0} x^{p-1} P[|\widehat{S}_n| > x] dx + \\ &\quad + p \sum_{k=1}^{\infty} \int_{4^k t_0}^{4^{k+1} t_0} x^{p-1} P[|\widehat{S}_n| > x] dx \\ &\leq (4t_0)^p + t_0^p \sum_{k=1}^{\infty} 4^{p(k+1)} P[|\widehat{S}_n| > 4^k t_0] \\ &\leq (4t_0)^p + 4^p t_0^p \sum_{k=1}^{\infty} 4^{kp} \gamma^{\sum_{i=1}^k 2^i} P^{2^k}[|\widehat{S}_n| > t_0] \\ &\leq (4t_0)^p + 4^p t_0^p \gamma^{-2} \sum_{k=1}^{\infty} 4^{kp} (\gamma^2 P[|\widehat{S}_n| > t_0])^{2^k} \\ &\leq (4t_0)^p + 4^p t_0^p \gamma^{-2} \sum_{k=1}^{\infty} \frac{4^{kp}}{4^{2^k}} \\ &= (4t_0)^p \left(1 + \frac{1}{\gamma^2}\right) \sum_{k=1}^{\infty} \frac{4^{kp}}{4^{2^k}} \\ &\leq 4^{\frac{p}{q}} 2 \gamma^{2p} 4^p \sum_{k=1}^{\infty} \frac{4^{kp}}{4^{2^k}} \max\{E^{\frac{p}{q}}|\widehat{S}_n|^q, c^p\} \\ &= B_{pq} \max\{E^{\frac{p}{q}}|\widehat{S}_n|^q, c^p\}. \end{aligned}$$

This is desired conclusion. \square

We will also need the following symmetrization result for slowly varying functions.

Lemma 6 *If $E[\widehat{X}^2 I_{[|\widehat{X}| \leq x]}]$ varies slowly then $E[X^2 I_{[|X| \leq x]}]$ varies slowly, too.*

PROOF OF LEMMA 6

Since $E|\widehat{X}| < \infty$ thus $E|X| < \infty$ (Cf. [25], p.243). By $P[|X| \geq 2E|X|] \leq \frac{1}{2}$ we have that $\text{median}(X) \leq 2E|X|$. Hence by the weak symmetrization inequalities (Cf. [25], p.245) we have

$$\frac{1}{2}P[|X| > x + 2E|X|] \leq P[|\widehat{X}| > x] \leq 2P[|X| > \frac{1}{2}x].$$

Therefore, if $x \geq eE|X|$ then

$$\begin{aligned} \frac{x^2 P[|X| > x]}{2 \int_0^x y P[|X| > y] dy} &\leq \frac{16x^2}{(x - 2E|X|)^2} \frac{(x - 2E|X|)^2 P[|\widehat{X}| > x - 2E|X|]}{2 \int_0^{x-2E|X|} y P[|\widehat{X}| > y] dy} \\ &\quad \times \frac{\int_0^{x-2E|X|} y P[|\widehat{X}| > y] dy}{\int_0^{2x} y P[|\widehat{X}| > y] dy}. \end{aligned} \quad (2.13)$$

Since the fraction standing left of the formula number (2.13) is at most 1 thus by (1.3), (2.13) and Theorem 2, VIII, §9 in [8]

$$\frac{x^2 P[|X| > x]}{x^2 P[|X| > x] + E[X_1^2 I_{[|X| \leq x]}]} = \frac{x^2 P[|X| > x]}{2 \int_0^x y P[|X| > y] dy} \rightarrow 0.$$

This proves Lemma 6. \square

By Lemma 2.1 in [31] and Corollary in [15] the sequence $\{f(a_k)\}$ fulfills $\psi_n \leq \varrho^n$, for some $\varrho < 0.8$ and $\psi_1 \leq 2 \ln 2 - 1 < 0.39$. Therefore, by (2.7) and (2.8) we have $\varphi_1 < 1/2$, $\widehat{\varphi}_1 < 1/2$, $\psi'_1 > 0$, $\widehat{\psi}'_1 > 0$, $\psi_1^* < 2$, $\widehat{\psi}_1^* < 2$, $\widehat{\rho}_1 \leq \widehat{\psi}_1 < 1$.

In particular, by Lemma 1 and (2.10) we have for every $x \geq 0$ and $n \geq 1$

$$P[\max_{1 \leq k \leq n} |\widehat{f}(a_k)| > 2x] \leq P[\max_{1 \leq k \leq n} |\widehat{S}_k| > x] \leq 6P[|\widehat{S}_n| > x]. \quad (2.14)$$

3 Proofs

PROOF OF THEOREM 1

Without the loss of generality we may assume that $E[f^2(a_1)] = \infty$ (the case $E[f^2(a_1)] < \infty$ is covered by Theorem 3). Assume first that (1.2) holds. Set $X_{nk} = f(a_k) I_{[|f(a_k)| \leq c_n]}$. By Proposition 2 we obtain

$$E\left(\sum_{k=1}^n (X_{nk} - E[X_{n1}])\right)^2 \leq Dn \text{Var}[X_{n1}].$$

In view of this and Chebyshev's inequality we get for any $\epsilon > 0$

$$P\left[\left|\sum_{k=1}^n (f(a_k) - E[f(a_k) I_{[|f(a_k)| \leq c_n]}])\right| > \epsilon c_n\right]$$

$$\begin{aligned}
&\leq \mathbb{P}\left[\left|\sum_{k=1}^n f(a_k)I_{[|f(a_k)| > c_n]}\right| > \frac{\epsilon}{2}c_n\right] + \mathbb{P}\left[\left|\sum_{k=1}^n (X_{nk} - \mathbb{E}[X_{nk}])\right| > \frac{\epsilon}{2}c_n\right] \\
&\leq n\mathbb{P}[|f(a_1)| > c_n] + 8Dn\epsilon^{-2}c_n^{-2}\mathbb{E}[f(a_1)^2I_{[|f(a_1)| \leq c_n]}] \rightarrow 0.
\end{aligned}$$

Conversely, assume (1.1). By (2.14)

$$\mathbb{P}\left[\max_{1 \leq k \leq n} |\hat{f}(a_k)| > \epsilon c_n\right] \rightarrow 0$$

so that denoting $Y_{nk}(\epsilon) = \hat{f}(a_k)I_{[|\hat{f}(a_k)| \leq \epsilon c_n]}$ we get

$$c_n^{-1} \sum_{k=1}^n Y_{nk}(\epsilon) \rightarrow_{\mathbb{P}} 0, \quad \epsilon \in (0, 1).$$

Let $Z_n(\epsilon) = \sum_{k=1}^n Y_{nk}(\epsilon)$ then $c_n^{-1}\hat{Z}_n(\epsilon) \rightarrow_{\mathbb{P}} 0$. Thus by Lemma 1

$$c_n^{-1} \max_{1 \leq k \leq n} \sum_{m=1}^k \hat{Y}_{nm}(\epsilon) \rightarrow_{\mathbb{P}} 0$$

and by Lemma 2 (with $x = \epsilon$, $X_k = c_n^{-1}\hat{Y}_{nk}$) we get

$$\lim c_n^{-2}\mathbb{E}[\hat{Z}_n^2(\epsilon)] = 0.$$

Since $\mathbb{E}[\hat{f}^2(a_1)] = 2\mathbb{E}[f^2(a_1)] = \infty$ thus $\mathbb{E}^2|Y_{n1}(\epsilon)| = o(\mathbb{E}[Y_{n1}^2(\epsilon)])$ (Cf. (2.6.14) in [14]). Therefore by Proposition 2

$$\lim_n \frac{n}{c_n^2} \mathbb{E}[Y_{n1}^2(\epsilon)] = \lim_n \frac{n}{c_n^2} \text{Var}[Y_{n1}(\epsilon)] = \lim_n \frac{n}{2c_n^2} \text{Var}[\hat{Y}_{n1}(\epsilon)] = 0, \quad \epsilon \in (0, 1)$$

so that $n^{-1}c_n^2 \rightarrow \infty$. Further, by (1.3) and since $n\mathbb{P}[|\hat{f}(a_1)| > \epsilon c_n] \rightarrow 0$ hence

$$\int_0^1 n\mathbb{P}[|f(a_1) - \tilde{f}(a_1)| > \sqrt{x\epsilon}c_n]dx \rightarrow 0, \quad \epsilon \in (0, 1).$$

Now, by the weak symmetrization inequalities (Cf. [25], p.245) we have

$$\begin{aligned}
&\int_0^1 n\mathbb{P}[|f(a_1)| > \sqrt{x}c_n]dx \\
&\leq \int_0^1 n\mathbb{P}[|f(a_1) - \text{median}(f(a_1))| > \sqrt{x}\frac{c_n}{2}]dx \\
&\quad + \int_0^1 n\mathbb{P}[|\text{median}(f(a_1))| > \sqrt{x}\frac{c_n}{2}]dx \\
&\leq \int_0^1 2n\mathbb{P}[|f(a_1) - \tilde{f}(a_1)| > \sqrt{x}\frac{c_n}{2}]dx \\
&\quad + \int_0^1 n\mathbb{P}[|\text{median}(f(a_1))| > \sqrt{x}\frac{c_n}{2}]dx \\
&\leq \int_0^1 2n\mathbb{P}[|f(a_1) - \tilde{f}(a_1)| > \sqrt{x}\frac{c_n}{2}]dx + 4\frac{n}{c_n^2}(\text{median}(f(a_1)))^2 \rightarrow 0.
\end{aligned}$$

This proves Theorem 1.

PROOF OF THEOREM 2

By (1.5) there exists $N \in \mathbb{N}$ such that for $n > N$

$$\frac{n-1}{c_{n-1}^2}(c_n^2 - c_{n-1}^2) \geq c > 1.$$

Setting, if necessary $c_k^2 = (c+1)^{-N+k+1}c_{N-1}^2$ for $1 \leq k < N$, we may assume that

$$\frac{n-1}{c_{n-1}^2}(c_n^2 - c_{n-1}^2) \geq c, \quad n > 1, \quad c > 1, \quad (3.15)$$

or, equivalently

$$\frac{c_n^2}{n} - \frac{c_{n-1}^2}{n-1} \geq \frac{c-1}{c} \frac{c_n^2 - c_{n-1}^2}{n}, \quad n > 1, \quad c > 1. \quad (3.16)$$

Further, by (3.15) we have

$$\ln c_n^2 \geq \ln c_1^2 + \sum_{k=2}^n \left(\ln \left(1 + \frac{c}{k-1} \right) - \frac{c}{k-1} \right) + c \sum_{k=2}^n \frac{1}{k-1}$$

and since by the Taylor expansion of $\ln(1+x)$ the first sum on the right hand side is $O(1)$ so that $\lim_{n \rightarrow \infty} \frac{\ln \frac{c_n^2}{n}}{\ln n} \geq c-1 > 0$. By this, (3.15) and (3.16) we get $\frac{c_n^2}{n} \nearrow \infty$.

Using in (3.16) the convention $\frac{0}{0} = 0$ we have

$$\begin{aligned} & \frac{n}{c_n^2} \mathbb{E}[f(a_1)^2 I_{[|f(a_1)| \leq c_n]}] \\ & \leq \frac{n}{c_n^2} \sum_{k=1}^n c_k^2 \mathbb{P}[|f(a_1)| \in (c_{k-1}, c_k]] \\ & = \frac{n}{c_n^2} \sum_{k=1}^n \sum_{\nu=1}^k (c_\nu^2 - c_{\nu-1}^2) \mathbb{P}[|f(a_1)| \in (c_{k-1}, c_k]] \\ & = \frac{n}{c_n^2} \sum_{\nu=1}^n (c_\nu^2 - c_{\nu-1}^2) \sum_{k=\nu}^n \mathbb{P}[|f(a_1)| \in (c_{k-1}, c_k]] \\ & \leq \frac{n}{c_n^2} \sum_{\nu=1}^n \left(\frac{c_\nu^2 - c_{\nu-1}^2}{\nu} \right) \nu \mathbb{P}[|f(a_1)| > c_{\nu-1}] \\ & \leq \frac{c}{c-1} \frac{n}{c_n^2} \sum_{\nu=1}^n \left(\frac{c_\nu^2}{\nu} - \frac{c_{\nu-1}^2}{\nu-1} \right) \nu \mathbb{P}[|f(a_1)| > c_{\nu-1}] \rightarrow_n 0 \end{aligned}$$

by the Toeplitz lemma (Cf. [25], p.238). In view of Theorem 1 the relation (1.6) is proved.

PROOF OF THEOREM 3

Suppose that $E[f^2(a_1)I_{[|f(a_1)| \leq x]}]$ is slowly varying and $E[f(a_1)] = 0$. Define the sequence b_n as follows: if $E[f^2(a_1)] < \infty$ then $b_n = \sigma\sqrt{n}$ provided that

$$\infty > \sigma^2 = E[f^2(a_1)] + 2 \sum_{k=2}^{\infty} E[f(a_1)f(a_k)] > 0;$$

if $E[f^2(a_1)] = \infty$ then

$$b_n = \sup\{x > 0; x^{-2}E[|f(a_1)|^2 I_{[|f(a_1)| \leq x]}] \geq \frac{1}{n}\}.$$

Assume first that $E[f^2(a_1)] < \infty$. Thus by Theorem 18.5.2 in [14] and Proposition 2 we have $0 < \sigma^2 < \infty$ and

$$\mathcal{L}(b_n^{-1} \sum_{k=1}^n f(a_k)) \rightarrow \mathcal{N}(0, 1).$$

Now, assume $E[f^2(a_1)] = \infty$. Then $b_n^2 \sim nE[f^2(a_1)I_{[|f(a_1)| \leq b_n]}]$ and $b_n \rightarrow \infty$. Let $X_{nk} = f(a_k)I_{[|f(a_k)| \leq b_n]}$, $S_n = \sum_{k=1}^n X_{nk}$. By the results in [34] if the following conditions are satisfied

$$\lim_{n \rightarrow \infty} nP[|f(a_n)| > b_n] = 0, \quad (3.17)$$

$$\tau_n^2 = \text{Var}(\sum_{k=1}^n X_{nk}) \rightarrow \infty, \quad (3.18)$$

$$\lim_{n \rightarrow \infty} \tau_n^{-2} E[\max_{1 \leq k \leq n} (X_{nk} - E[X_{nk}])^2] = 0 \quad (3.19)$$

then

$$\mathcal{L}(\tau_n^{-1}(S_n - nE[X_{n1}])) \rightarrow \mathcal{N}(0, 1). \quad (3.20)$$

The condition (3.17) easily follows by the definition of b_n and the slow variation of $E[f^2(a_1)I_{[|f(a_1)| \leq x]}]$ (Cf. Theorem 2, VIII, §9 in [8]). For (3.18) let us observe that by (2.9) and since $E[f^2(a_1)] = \infty$ thus $E^2[X_{n1}] = o(E[X_{n1}^2])$ (Cf. (2.6.14) in [14]) and therefore we have $b_n^2 \sim n\text{Var}[X_{n1}]$. Now by

$$\begin{aligned} \tau_n^2 &= E(\sum_{k=1}^n (X_{nk} - E[X_{n1}]))^2 \\ &= n\text{Var}[X_{n1}] + 2 \sum_{k=2}^n (n - k + 1) \text{Cov}[X_{n1}X_{nk}] \\ &= n\text{Var}[X_{n1}](1 + O(\frac{2}{(1 - \varrho)} \frac{n}{b_n^2} E^2[X_{n1}])) = b_n^2(1 + o(1)) \end{aligned}$$

(3.18) follows. For the relation (3.19) note that for every $\epsilon \in (0, 1)$, some $K > 0$ independent of n by the slow variation of $E[f^2(a_1)I_{[|f(a_1)| \leq x]}]$

$$\begin{aligned} &\tau_n^{-2} E[\max_{1 \leq k \leq n} (X_{nk} - E[X_{nk}])^2] \\ &\leq K(b_n^{-2} n E[f^2(a_1)I_{[|f(a_1)| \leq b_n]}] E[f^2(a_1)I_{[|f(a_1)| \leq b_n] > \epsilon b_n^2}]) + \epsilon \\ &\leq K(b_n^{-2} n E[f^2(a_1)I_{[|f(a_1)| \in (\sqrt{\epsilon} b_n, b_n]}]) + \epsilon \leq K(o(1) + \epsilon). \end{aligned}$$

Thus (3.20) holds and since $E[f(a_1)] = 0$, $\tau_n \sim b_n$ so that $\mathcal{L}(b_n^{-1}S_n) \rightarrow \mathcal{N}(0, 1)$. If $E[f(a_1)] \neq 0$ then we consider the sequence $\{f(a_k) - E[f(a_1)]\}$. It is worth noting that by Theorem 1.9.8 in [3] the sequence $\{b_n\}$ can be replaced by $\{c_n\}$ such that $\lim_n \frac{n}{c_n^2}(c_{n+1}^2 - c_n^2) = 1$.

Conversely, assume $\phi_n(c_n^{-1}\theta)e^{-i\theta\frac{dn}{c_n}} = E[e^{i\theta c_n^{-1}\sum_{k=1}^n f(a_k)}]e^{-i\theta\frac{dn}{c_n}} \rightarrow e^{-\frac{\theta^2}{2}}$ thus we have $\hat{\phi}_n(c_n^{-1}\theta) = E[e^{i\theta c_n^{-1}\sum_{k=1}^n \hat{f}(a_k)}] \rightarrow e^{-\theta^2}$. By Theorem 3.1 in [18] we have $c_n^2 = nh(n)$, where $h(n)$ is a slowly varying sequence. There is no loss of generality in assuming that $E[f^2(a_1)] = \infty$ so that $\sup\{x; P[|\hat{f}(a_1)| \leq x] < 1\} = \infty$. By Lemma 4 we have $\sup_n c_n^{-q} E|\sum_{k=1}^n \hat{f}(a_k)|^q < \infty$, $q \in (0, 2)$, so by (2.14) we obtain that the sequence $\{c_n^{-1} \max_{1 \leq k \leq n} |\hat{f}(a_k)|\}$ is uniformly integrable. On the other hand from Lemma 3 we get that

$$\begin{aligned} & (1 - \hat{\varphi}_1)\hat{\psi}_1' \lim_n (1 - e^{-nP[\hat{f}(a_1) > \epsilon c_n]}) \\ & \leq 4m \lim_n P[|\hat{S}_{[n/m]+1}| > \frac{\epsilon}{2}c_n] = \frac{4m}{\sqrt{\pi}} \int_{\frac{\epsilon}{2}\sqrt{m}}^{\infty} e^{-u^2/4} du \end{aligned}$$

and letting $m \rightarrow \infty$ it yields $\{c_n^{-1} \max_{1 \leq k \leq n} |\hat{f}(a_k)|\} \rightarrow 0$ in probability since the integral tends to 0 faster than exponentially. Moreover, by Theorem 3.5 in [2] the latter convergence takes place in L^1 , too. Now, by Proposition 3 (with $q = 1, x = \epsilon, X_k = c_n^{-1}\hat{f}(a_k)$) we obtain

$$\frac{n}{c_n} E[|\hat{f}(a_1)| I_{[|\hat{f}(a_1)| > \epsilon c_n]}] \rightarrow 0.$$

Hence using the notation from the proof of Theorem 1 we get

$$\begin{aligned} & \sup_n c_n^{-1} E|Z_n(\epsilon)| \\ & \leq \sup_n c_n^{-1} E|\sum_{k=1}^n \hat{f}(a_k)| + \sup_n c_n^{-1} E|\sum_{k=1}^n \hat{f}(a_k) I_{[|\hat{f}(a_k)| > \epsilon c_n]}| \\ & \leq \sup_n c_n^{-1} E|\sum_{k=1}^n \hat{f}(a_k)| + \sup_n \frac{n}{c_n} E[|\hat{f}(a_1)| I_{[|\hat{f}(a_1)| > \epsilon c_n]}] < \infty. \end{aligned}$$

Consequently $\sup_n c_n^{-1} E|\hat{Z}_n(\epsilon)| < \infty$ and Lemma 5 (with $p = 2, q = 1$) gives $\sup_n c_n^{-2} E[\hat{Z}_n^2(\epsilon)] < \infty$. In view of this it follows from Proposition 2 that $\sup_n \frac{n}{c_n^2} E[Y_{n1}^2(\epsilon)] < \infty$. Therefore there have to be $n^{-1}c_n^2 \rightarrow \infty$ since we assumed $E[f^2(a_1)] = \infty$.

By Lemma 1 in [11] for $\theta \in \mathbb{R}$ and any integers $m > 0, p > 1$ such that

$$E|e^{i\theta c_n^{-1}\hat{f}(a_1)} - 1| \leq \min\left\{\frac{1}{2(1 + \hat{\psi}_1)^2(2m + 1)^2}, \frac{1}{2(1 + \hat{\psi}_1)(2pm + 1)}\right\} \quad (3.21)$$

and

$$nE^2|e^{i\theta c_n^{-1}\hat{f}(a_1)} - 1| \leq \frac{1}{2}\left(9 + \sum_{\nu=1}^m \hat{\psi}_\nu\right)^{-1} n(1 - \hat{\phi}_1(c_n^{-1}\theta)) \quad (3.22)$$

we have

$$\begin{aligned}
& |\widehat{\phi}_n(c_n^{-1}\theta) - \exp\{n(\widehat{\phi}_1(c_n^{-1}\theta) - 1)\}| \\
& \leq (9 + \sum_{\nu=1}^m \widehat{\psi}_\nu) n E^2 |e^{i\theta c_n^{-1}\widehat{f}(a_1)} - 1| \exp\{-\frac{1}{2}n(1 - \widehat{\phi}_1(c_n^{-1}\theta))\} \\
& \quad + (2^{-p} + (6 + \widehat{\psi}_1)\widehat{\psi}_{m+1}) n E |e^{i\theta c_n^{-1}\widehat{f}(a_1)} - 1|.
\end{aligned} \tag{3.23}$$

Now, observe that

$$\sqrt{n} E |e^{i\theta c_n^{-1}\widehat{f}(a_1)} - 1| \leq \sqrt{\frac{n}{c_n^2}} |\theta| E |\widehat{f}(a_1)| = o(1). \tag{3.24}$$

By (3.24) we can put $m = p \equiv \lfloor \sqrt[4]{n} \rfloor$ in (3.21) and (3.22) so that by (3.23) we get

$$\overline{\lim}_n |\widehat{\phi}_n(c_n^{-1}\theta) - \exp\{n(\widehat{\phi}_1(c_n^{-1}\theta) - 1)\}| = 0.$$

Since $\widehat{\phi}_n(c_n^{-1}\theta) \rightarrow e^{-\theta^2}$ thus $n(\widehat{\phi}_1(c_n^{-1}\theta) - 1) \rightarrow -\theta^2$. Whence by the proof of Theorem 8.3.1 in [3] we have that $E[\widehat{f}^2(a_1)I_{[|\widehat{f}(a_1)| \leq x]}]$ is a slowly varying function in the sense of Karamata and by Lemma 6 we get that $E[f^2(a_1)I_{[|f(a_1)| \leq x]}]$ varies slowly, too. Further, by the direct part of this proof we know that one can choose $d_n = nE[f(a_1)]$, which is finite under the slow variation condition. This completes the proof of Theorem 3.

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